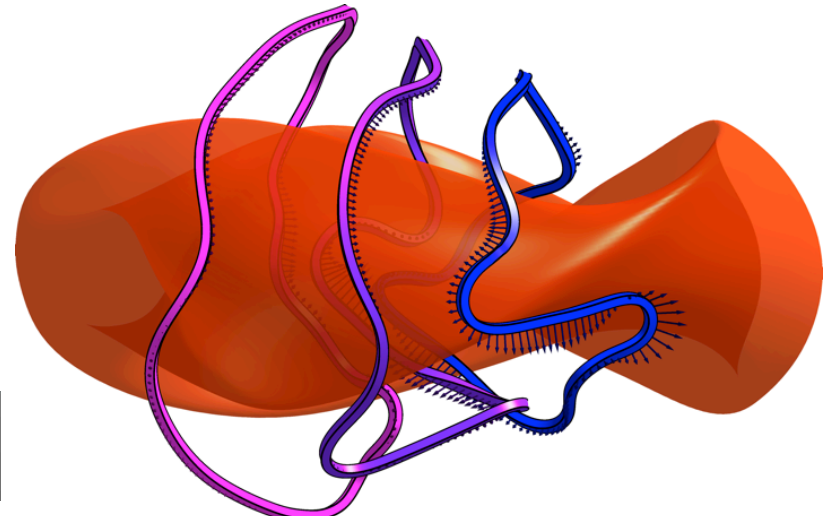
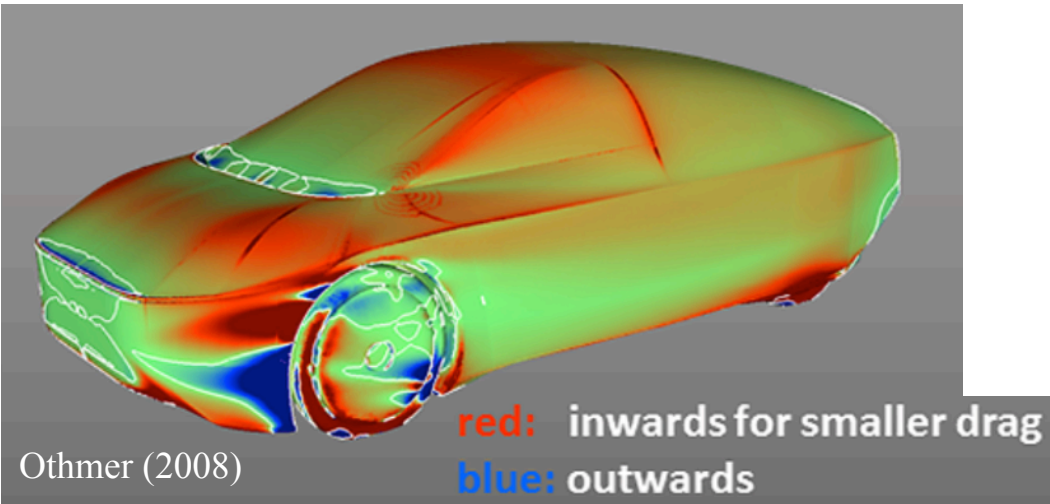


# Understanding local sensitivity & tolerances of stellarators using shape gradients

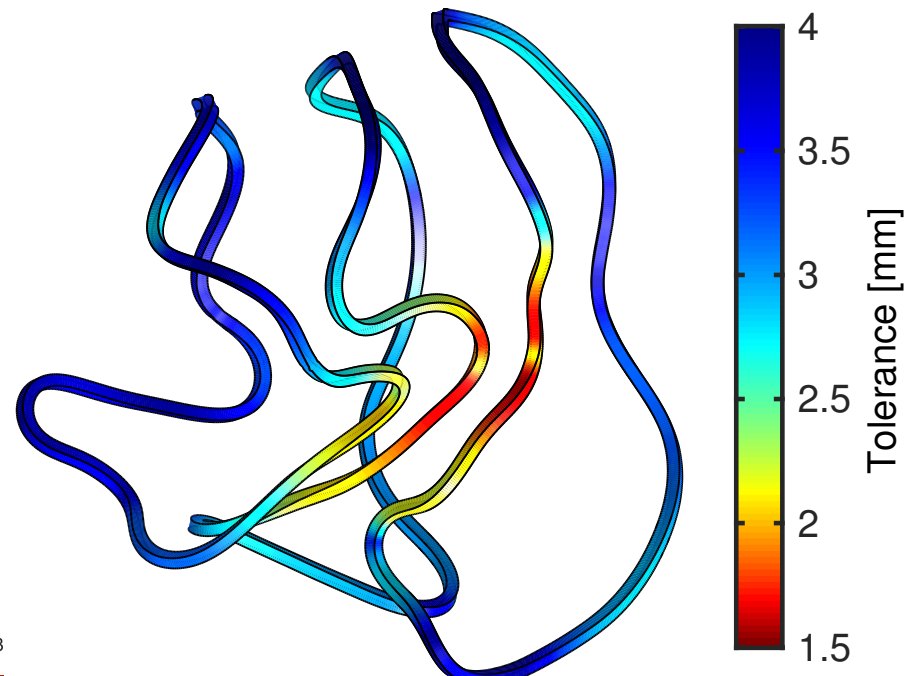
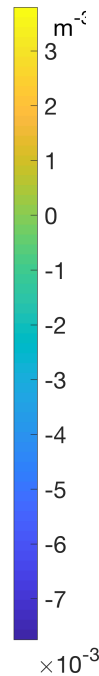
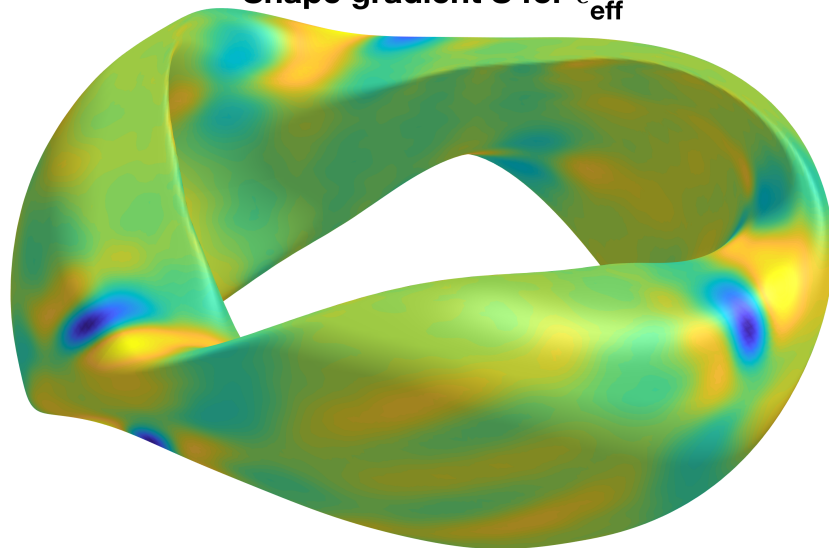
Matt Landreman, Elizabeth Paul,

U of Maryland

Nuclear Fusion 58, 076023 (2018)



Shape gradient  $S$  for  $\epsilon_{\text{eff}}^{3/2}$



Tolerance [mm]

# The shape gradient is a new (to us) way to think about derivatives involving shapes.

- Derivatives involving shapes are central to stellarator optimization.
- These derivatives also encode tolerances, which have been a leading driver of cost:

“The largest driver of the project cost growth were the accuracy requirements.” [Strykowski et al, *Engineering Cost & Schedule Lessons Learned on NCSX*, (2009)]

- Compared to ‘parameter derivatives’, shape gradients have 2 advantages:
  - *Spatially local*
  - *Independent of parameterization*

# Understanding local sensitivity of stellarators using shape gradients

- 2 ways to represent derivatives: parameter derivatives vs. shape gradients.
- Computing shape gradients from parameter derivatives (e.g. STELLOPT)
- Examples
  - Rotational transform
  - Neoclassical transport
  - Coil-plasma distance
- Coil tolerances
- Magnetic sensitivity and tolerances

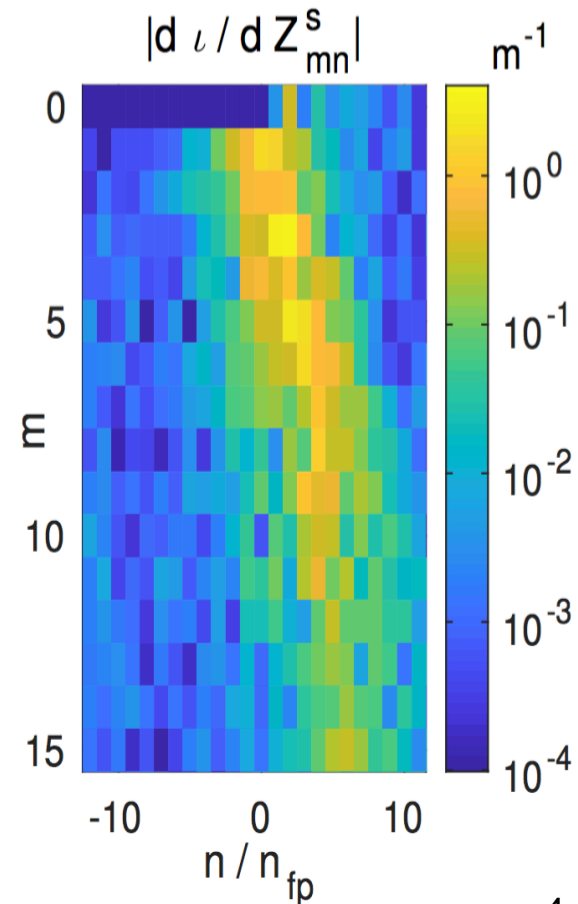
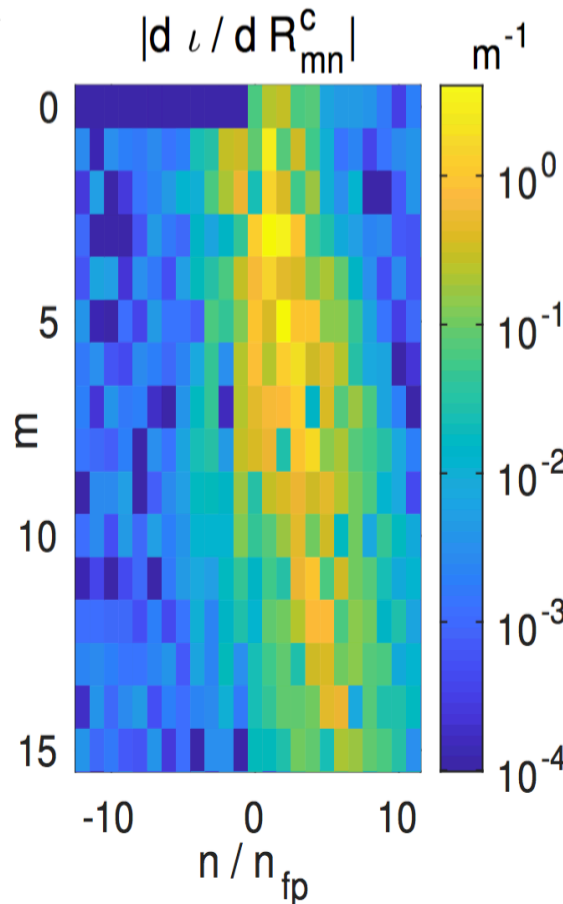
# Parameter derivatives and shape gradients are complementary ways to express derivatives.

Let  $f$  denote any figure of merit, e.g.  $\iota$ , neoclassical transport, etc.

**Parameter derivatives:** Example:  $\partial f / \partial R_{m,n}^c$  and  $\partial f / \partial Z_{m,n}^s$   
where  $R_{m,n}^c$  and  $Z_{m,n}^s$  parameterize the plasma boundary shape:

$$R(\theta, \zeta) = \sum_{m,n} R_{m,n}^c \cos(m\theta - n\zeta),$$

$$Z(\theta, \zeta) = \sum_{m,n} Z_{m,n}^s \sin(m\theta - n\zeta)$$





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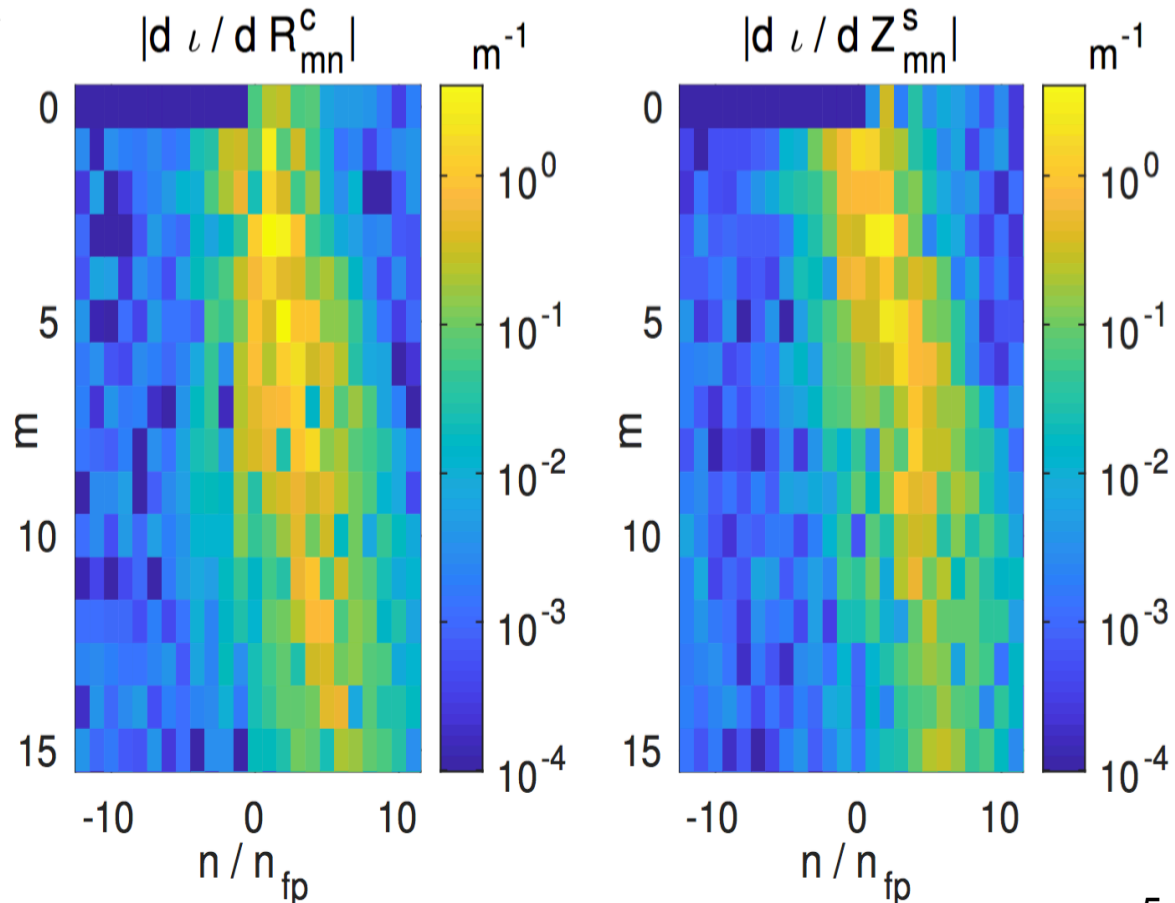
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$$Z(\theta, \zeta) = \sum_{m,n} Z_{m,n}^s \sin(m\theta - n\zeta)$$

- Successfully used in STELLOPT to design NCSX, etc.
- Computable by finite differencing any code.

But,

- Not unique: coordinate-dependent,
- Nonlocal: awkward for engineering.

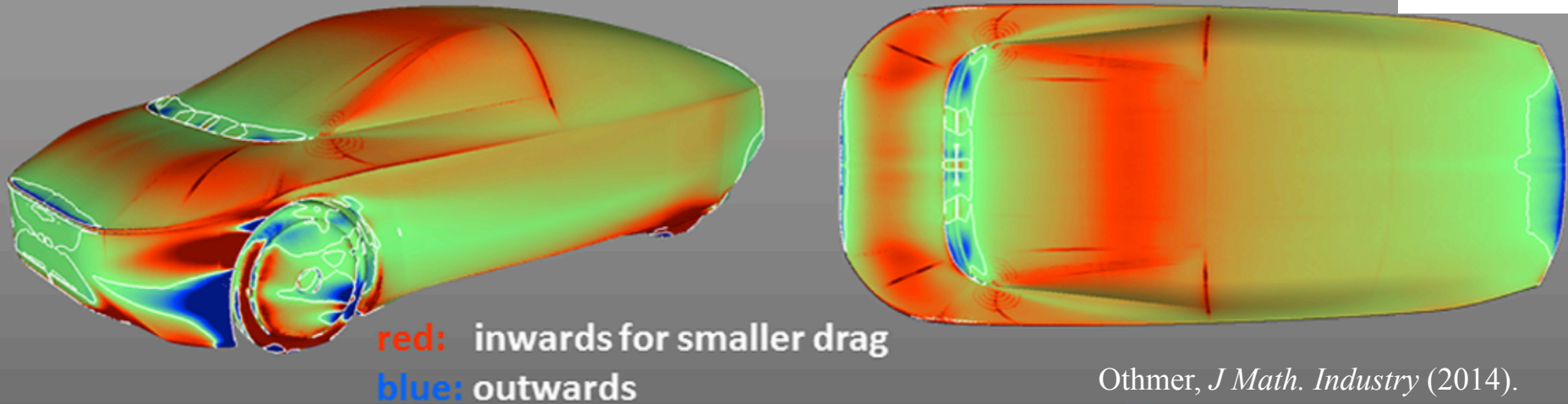


# Parameter derivatives and shape gradients are complementary ways to express derivatives.

## Shape gradients:

For surfaces,  $S$  where  $\delta f = \int d^2a (\delta \mathbf{r} \cdot \mathbf{n}) S$

Unit normal



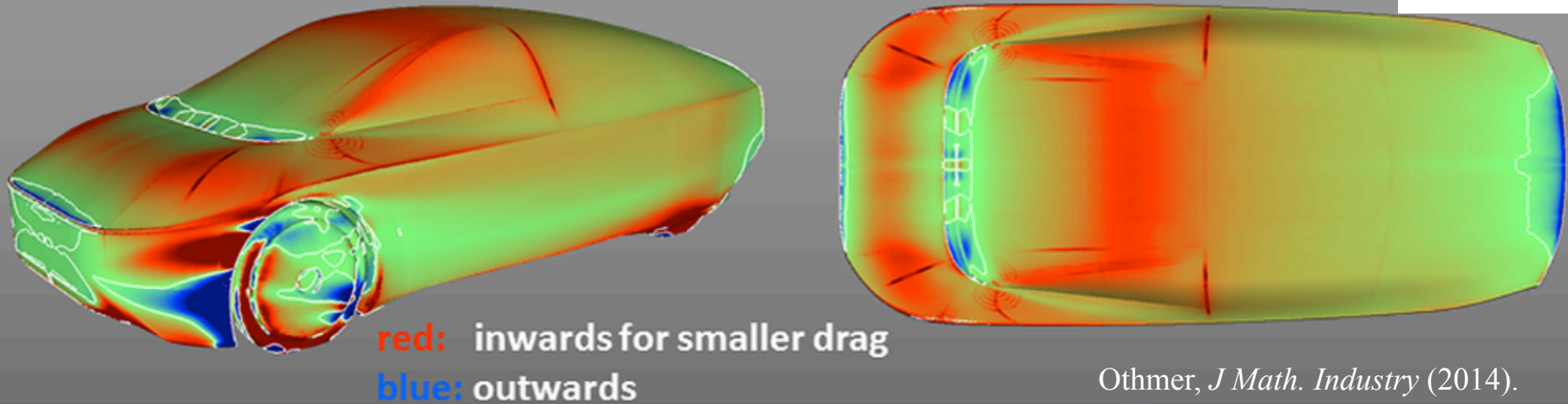
- Local (real-space, not Fourier-space). More useful for engineering.
- Independent of coordinates used to parameterize surface.
- May provide different insight than parameter derivatives.

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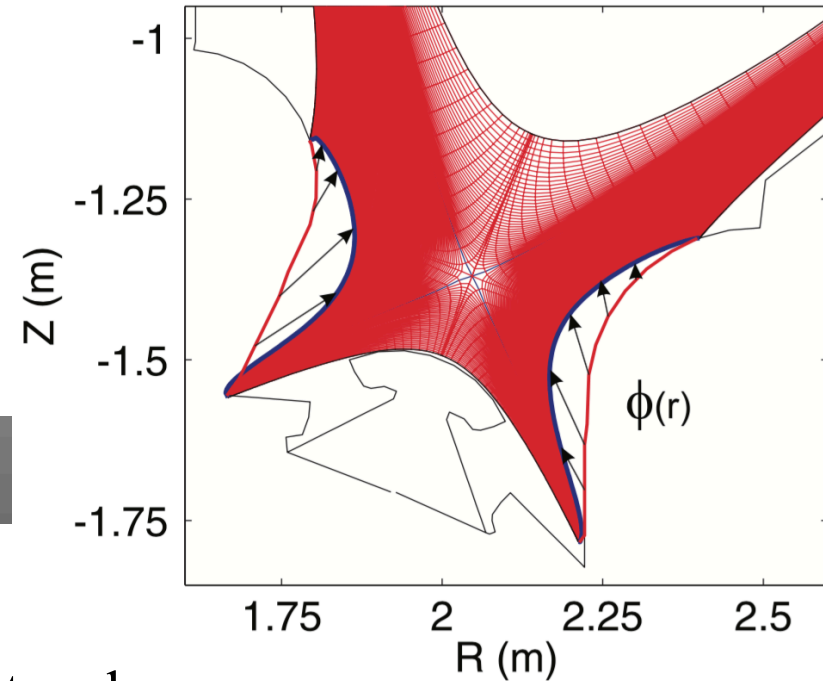
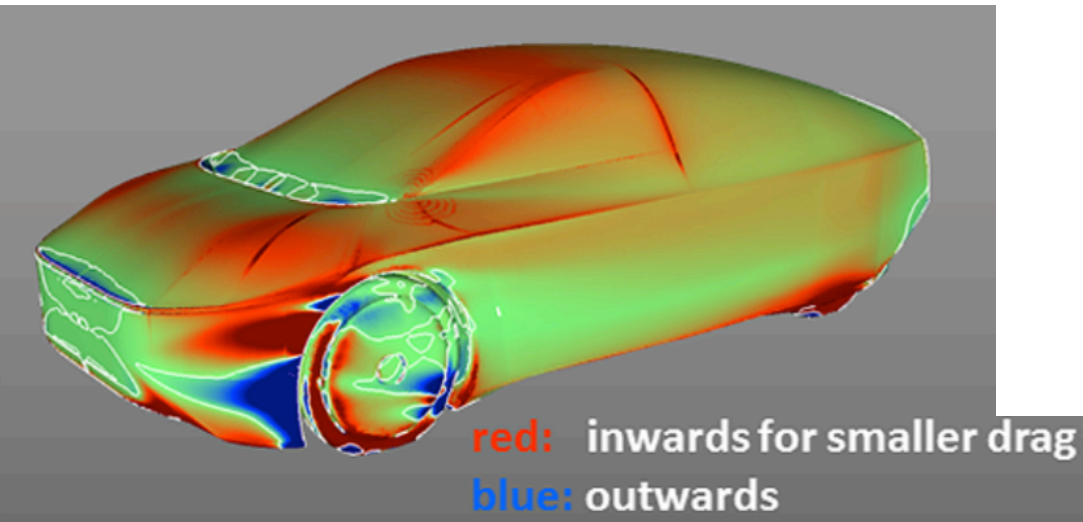
- Local (real-space, not Fourier-space). More useful for engineering.
- Independent of coordinates used to parameterize surface.
- May provide different insight than parameter derivatives.

For coils,  $\mathbf{S}_k$  where  $\delta f = \sum_{\text{coils } k} \int d\ell \delta \mathbf{r} \cdot \mathbf{S}_k$ , and  $\mathbf{S}_k \cdot \frac{\partial \mathbf{r}}{\partial \ell} = 0$ .

# Parameter derivatives and shape gradients are complementary ways to express derivatives.

## Shape gradients:

For surfaces,  $S$  where  $\delta f = \int d^2a (\delta \mathbf{r} \cdot \mathbf{n}) S$



Recently used for optimizing tokamak divertor shapes:

- W. Dekeyser, Ph.D. thesis, KU Leuven (2014).
- W. Dekeyser et al, *Nucl. Fusion* 54, 073022 (2014).
- M. Baelmans, et al, *Nucl. Fusion* 57, 036022 (2017).

# The shape gradient representation can be expected to exist for many important shape functionals.

1D:

Derivative of a function of  $n$  numbers  $f(r_1, r_2, \dots, r_n)$ :

$$\delta f = \sum_{j=1}^n \frac{\partial f}{\partial r_j} \delta r_j$$

$$n \rightarrow \infty \text{ limit: } f = f[r(\ell)], \quad \delta f = \int d\ell \underbrace{\frac{\delta f}{\delta r}}_S \delta r$$

This is an instance of the Riesz representation theorem:

*Any linear operator can be written as an inner product with some element of the appropriate space.*

# In some cases, shape gradients can be computed analytically.

## Integrals over a curve:

If  $f[C] = \int_C d\ell Q$  for some  $Q(\mathbf{r})$  and space curve  $C$ ,

$$\Rightarrow \delta f = \int_C d\ell \delta \mathbf{r} \cdot \underbrace{\left[ (\vec{\mathbf{I}} - \mathbf{t}\mathbf{t}) \cdot \nabla Q - Q\kappa \mathbf{n} \right]}_S$$

where  $\kappa$  = curvature,  $\mathbf{t}$  = tangent.

## Integrals over a surface:

If  $f[\partial\Omega] = \int_{\partial\Omega} d^2a Q$  for some  $Q(\mathbf{r})$  and surface  $\partial\Omega$ ,

$$\Rightarrow \delta f = \int_{\partial\Omega} d^2a (\delta \mathbf{r} \cdot \mathbf{n}) \underbrace{(\mathbf{n} \cdot \nabla Q - 2QH)}_S$$

where  $H$  = mean curvature.

# Understanding local sensitivity of stellarators using shape gradients

- 2 ways to represent derivatives: parameter derivatives vs. shape gradients.
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# Previously, shape gradients have usually been computed using adjoint methods.

- An adjoint equation, when available, gives the shape gradient (representing all possible shape perturbations) at the cost of  $\sim 1$  regular solve.
- Then, the shape gradient is very efficient for shape optimization:

$$\text{Choose } \delta \mathbf{r} = -\varepsilon S \mathbf{n} \quad \Rightarrow \quad \delta f = -\varepsilon \int d^2 a S^2 < 0.$$

$\nwarrow$  Step size  $\varepsilon > 0$

- However, adjoint methods require analytic work & code development for every target.
- Here, instead, we show how to compute shape gradients from any code with little extra work.

**The shape gradient can be computed from parameter derivatives by solving a small linear system.**

**Coils:** Discretize coil shapes:

$$X(\vartheta) = X_0^c + \sum_{m=1} \left[ X_m^c \cos(m\vartheta) + X_m^s \sin(m\vartheta) \right] \quad \& \ Y, Z$$

Parameters  $p_j$  are  $\{X_m^c, X_m^s, Y_m^c, Y_m^s, Z_m^c, Z_m^s\}$ .

Compute  $\partial f / \partial p_j$  using finite differences, e.g. STELLOPT.

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Discretize shape gradient:

$$S_X(\vartheta) = S_{X,0}^c + \sum_{m=1} \left[ S_{X,m}^c \cos(m\vartheta) + S_{X,m}^s \sin(m\vartheta) \right] \quad \& S_Y, S_Z$$

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$$\int d\ell \, \delta \mathbf{r} \cdot \mathbf{S} = \delta f \quad \Rightarrow \quad \text{Solve } \int d\ell \, \frac{\partial \mathbf{r}}{\partial p_j} \cdot \mathbf{S} = \frac{\partial f}{\partial p_j} \text{ for } \mathbf{S}.$$

(Square linear system)

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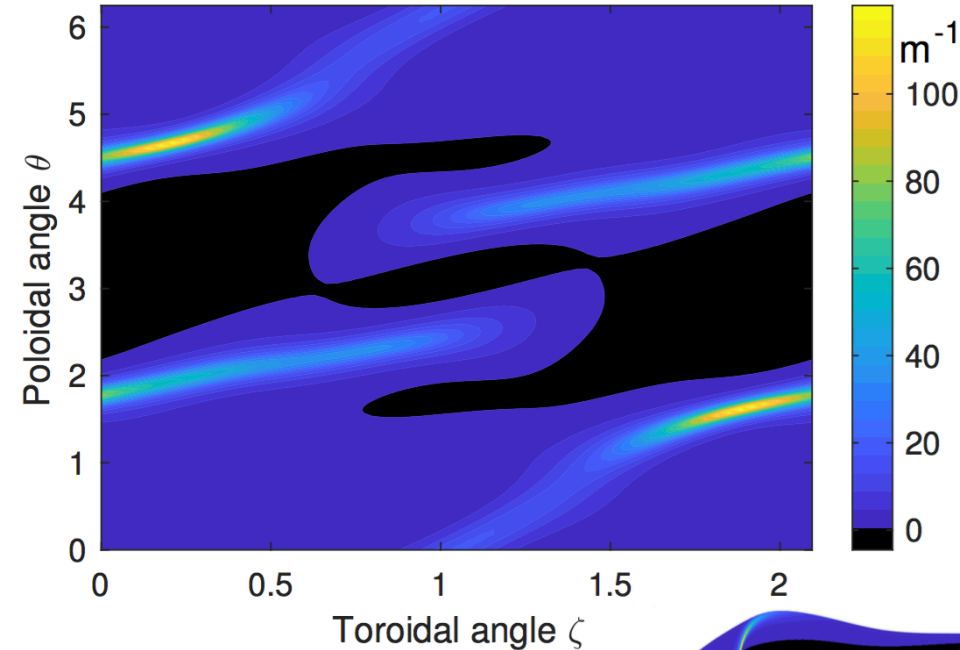
$$\int d\ell \, \delta \mathbf{r} \cdot \mathbf{S} = \delta f \quad \Rightarrow \quad \text{Solve } \int d\ell \, \frac{\partial \mathbf{r}}{\partial p_j} \cdot \mathbf{S} = \frac{\partial f}{\partial p_j} \text{ for } \mathbf{S}.$$

Similar procedure for boundary surface.

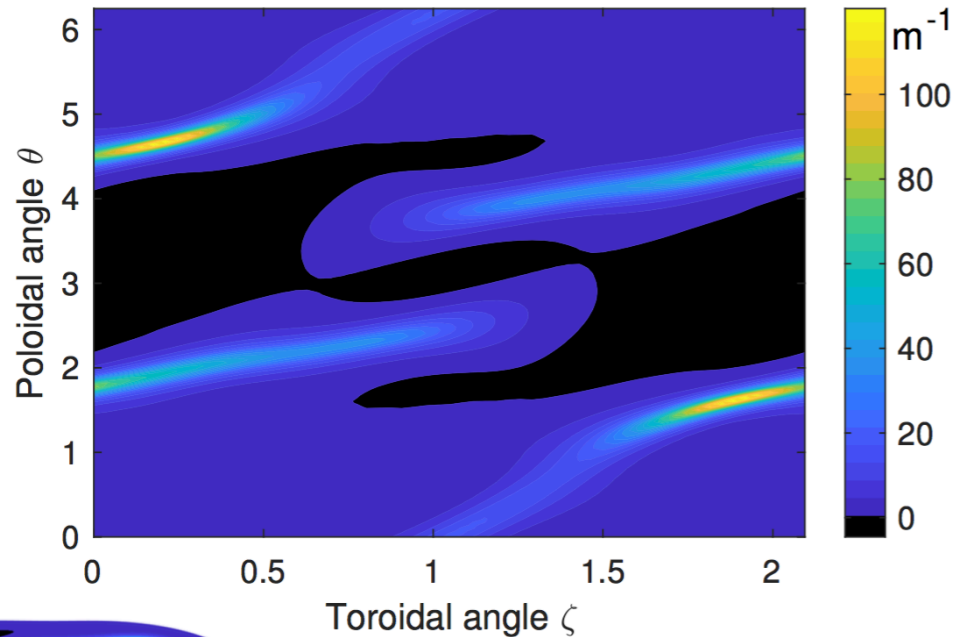
# The algorithm for computing shape gradients can be verified by comparison to analytic theory.

Consider  $f = \text{area}$ . Analytic result:  $S = -2 \times (\text{mean curvature})$

Shape gradient  $S$  computed from  $-2 \times (\text{mean curvature})$



Shape gradient  $S$  computed using  $d/dR_{mn}$  &  $d/dZ_{mn}$



$$\delta f = \int d^2 a (\delta \mathbf{r} \cdot \mathbf{n}) S$$

# Understanding local sensitivity of stellarators using shape gradients

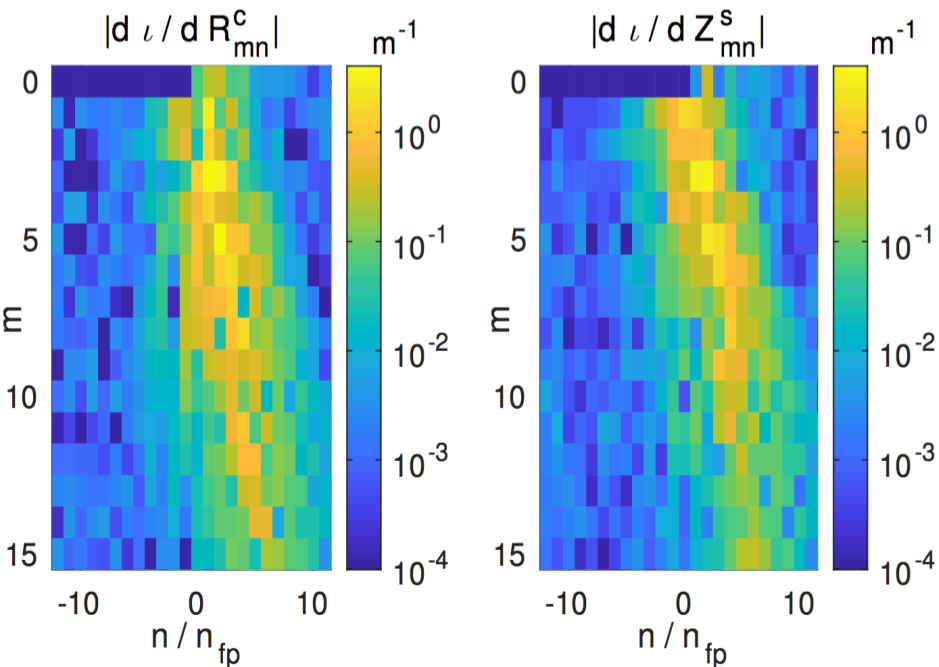
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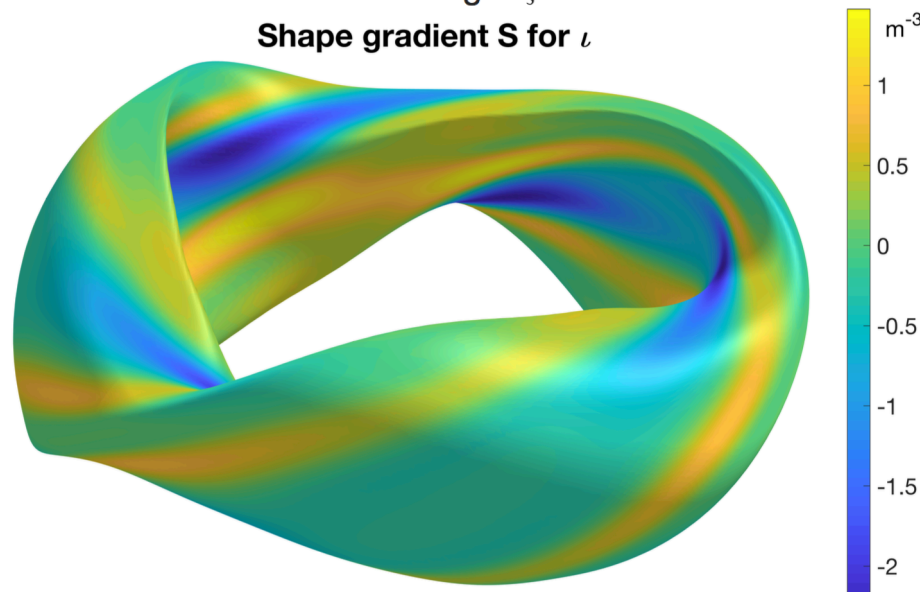
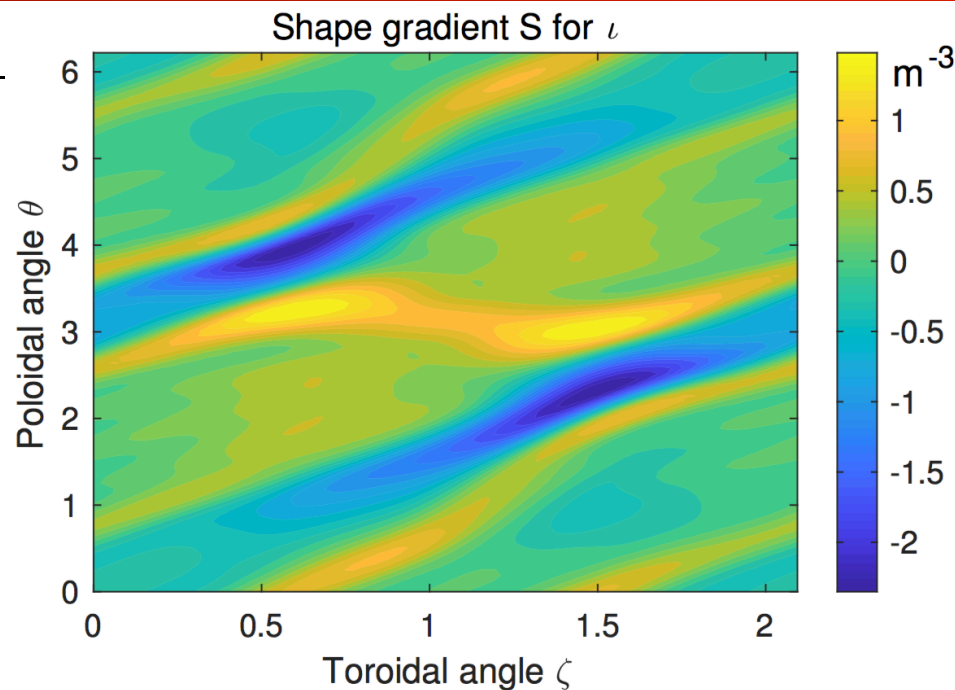
# Example: Rotational transform at $r/a=0.5$

## Shape gradient for boundary surface:

Parameter derivatives from  
STELLOPT/VMEC:

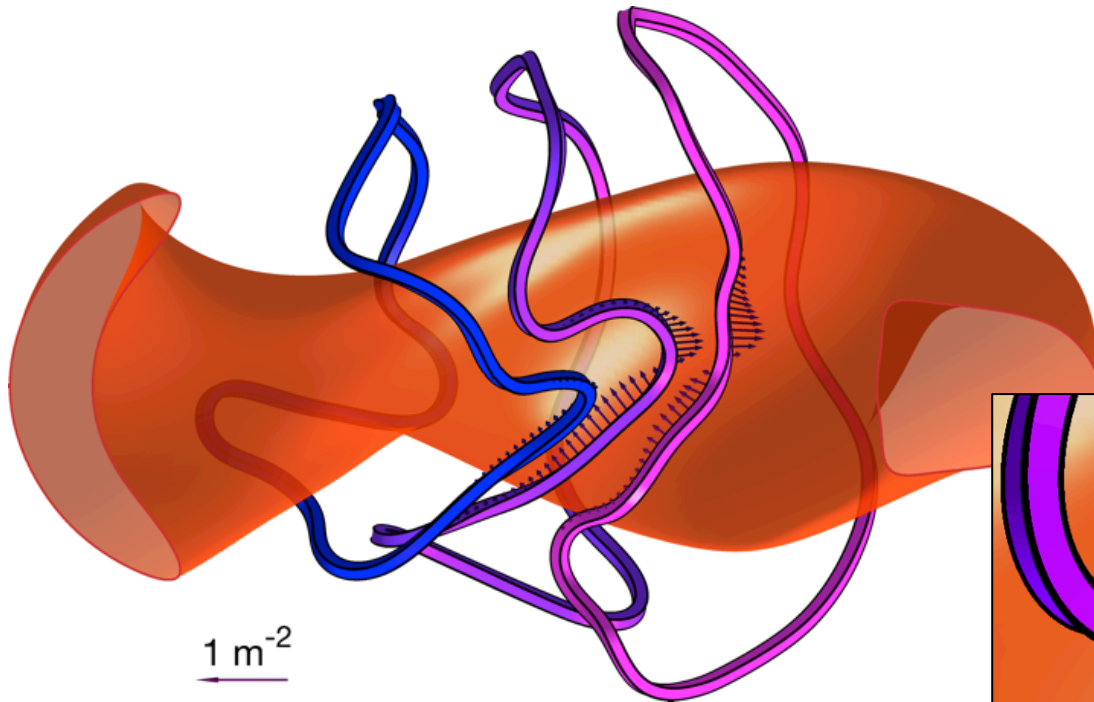


$$\delta f = \int d^2a (\delta \mathbf{r} \cdot \mathbf{n}) S$$



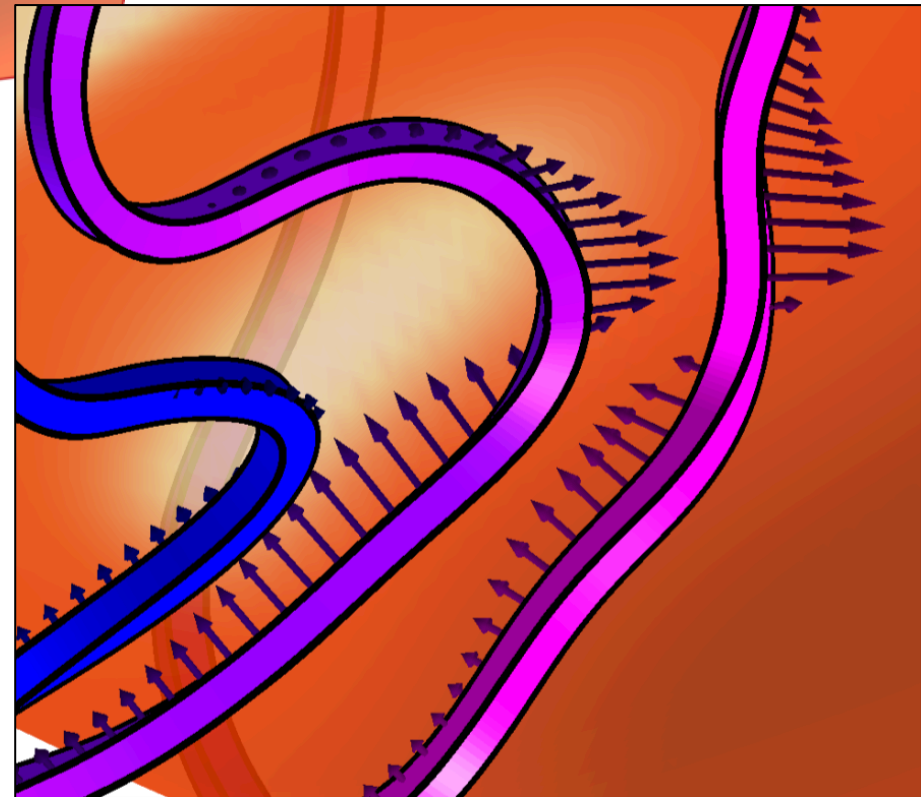
# Example: Rotational transform at $r/a=0.5$

Shape gradient for coils:



Arrows show  $\mathbf{S}_k$  where

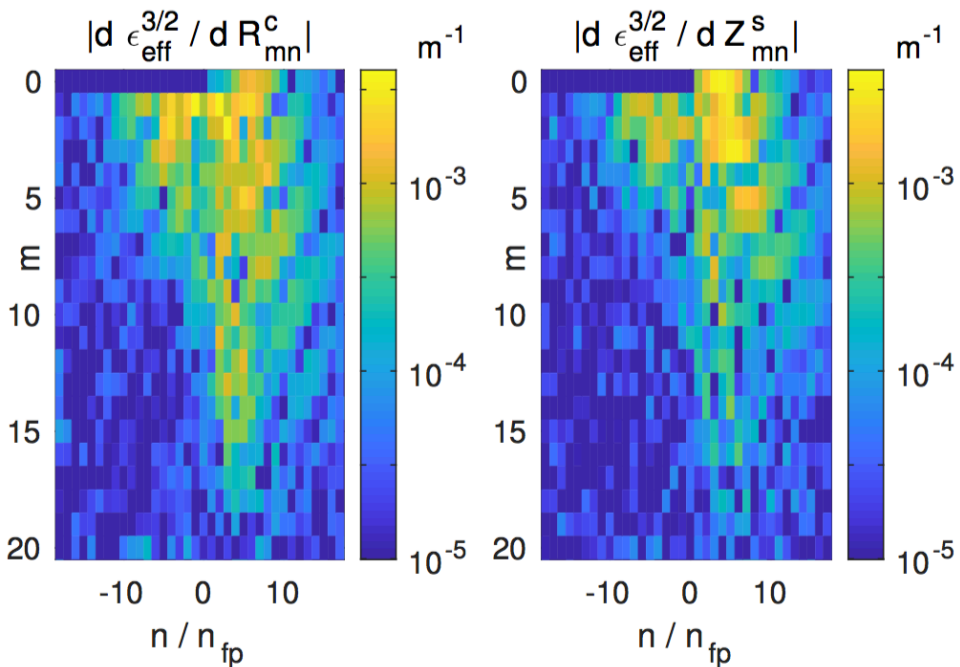
$$\delta f = \sum_{\text{coils } k} \int d\ell \, \delta \mathbf{r} \cdot \mathbf{S}_k.$$



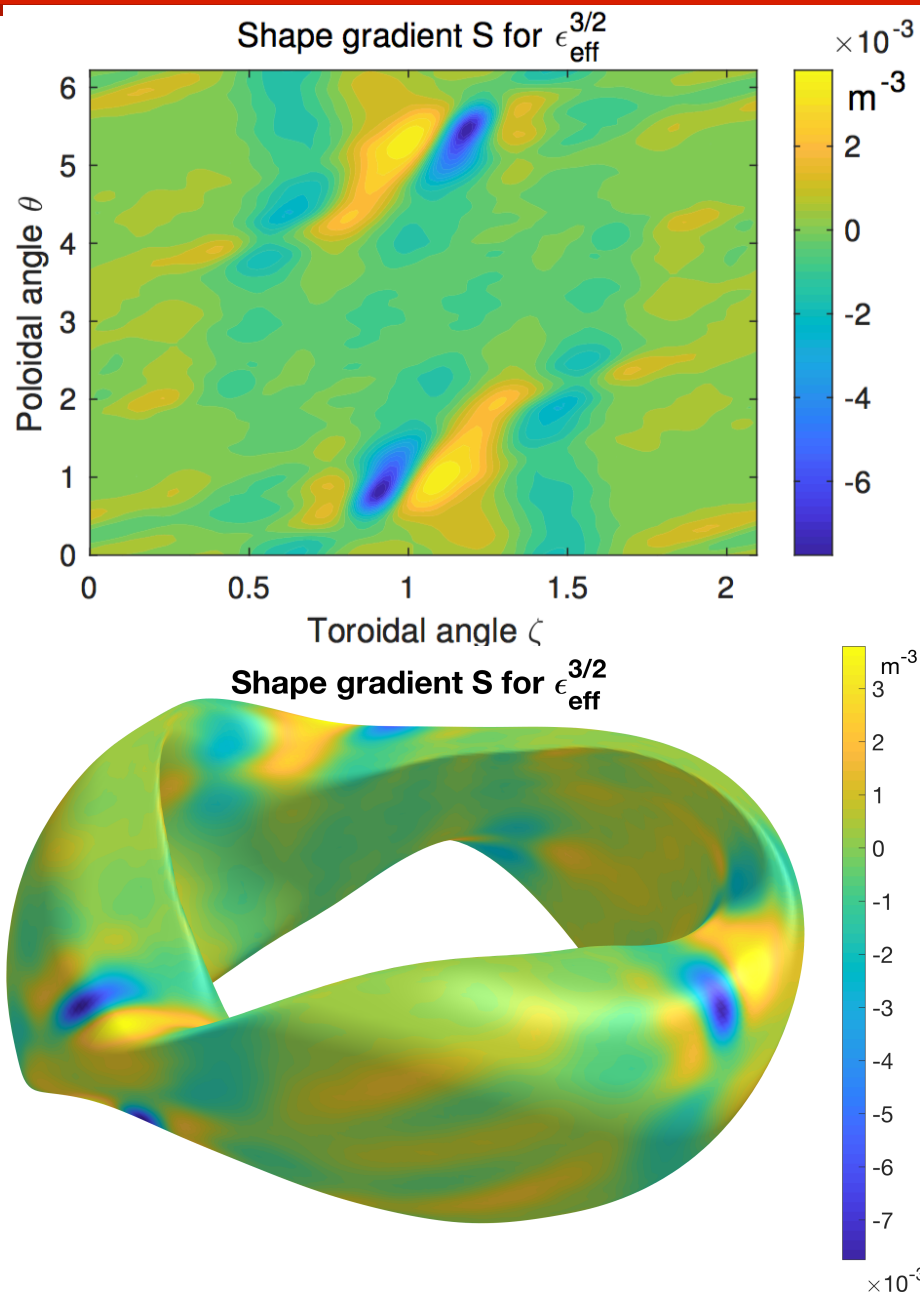
# Example: Neoclassical transport $\epsilon_{\text{eff}}^{3/2}$ at $r/a=0.5$

## Shape gradient for boundary surface:

Parameter derivatives from  
STELLOPT/VMEC/NEO:

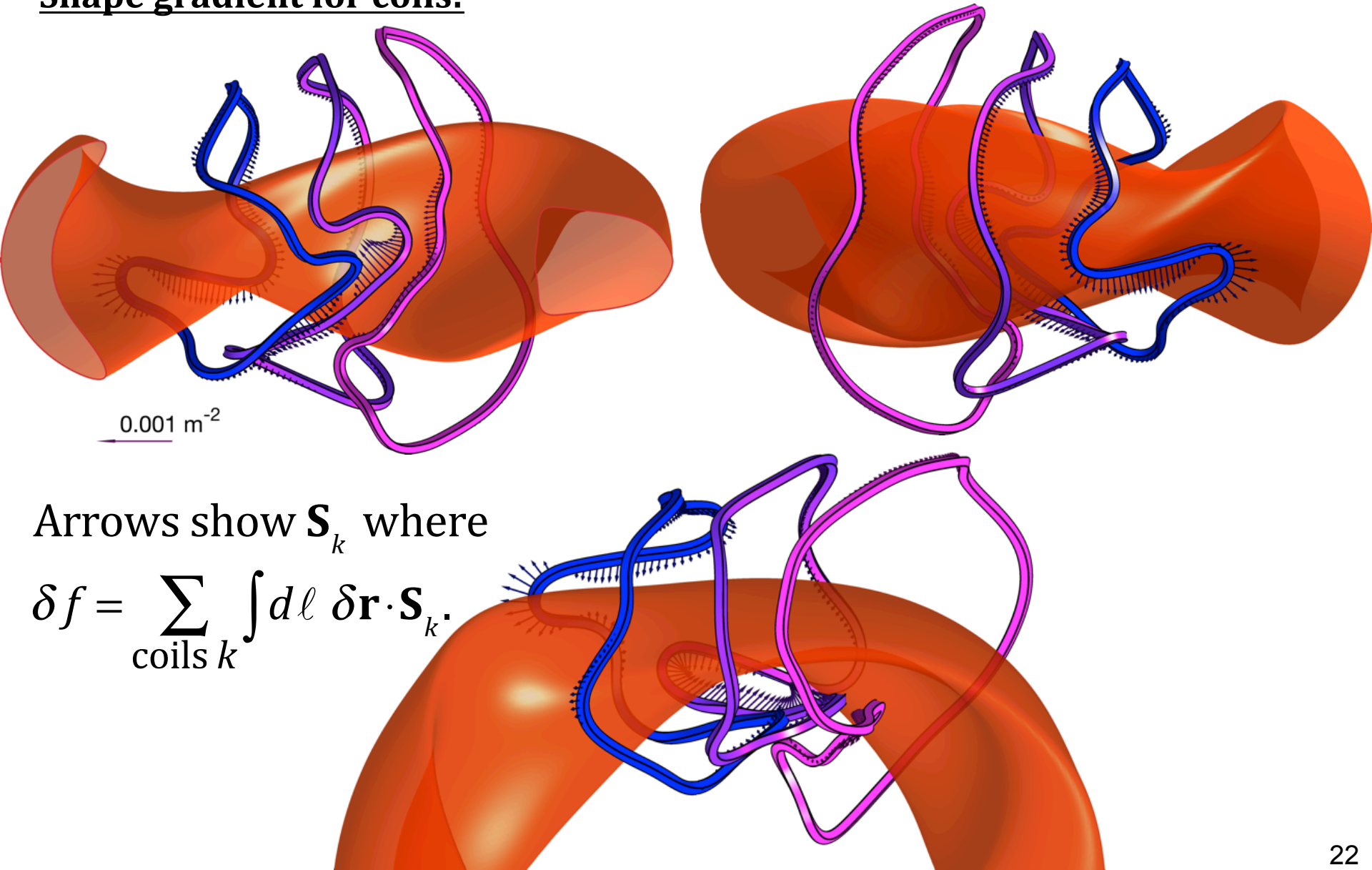


$$\delta f = \int d^2 a (\delta \mathbf{r} \cdot \mathbf{n}) S$$



# Example: Neoclassical transport $\varepsilon_{\text{eff}}^{3/2}$ at $r/a=0.5$

Shape gradient for coils:



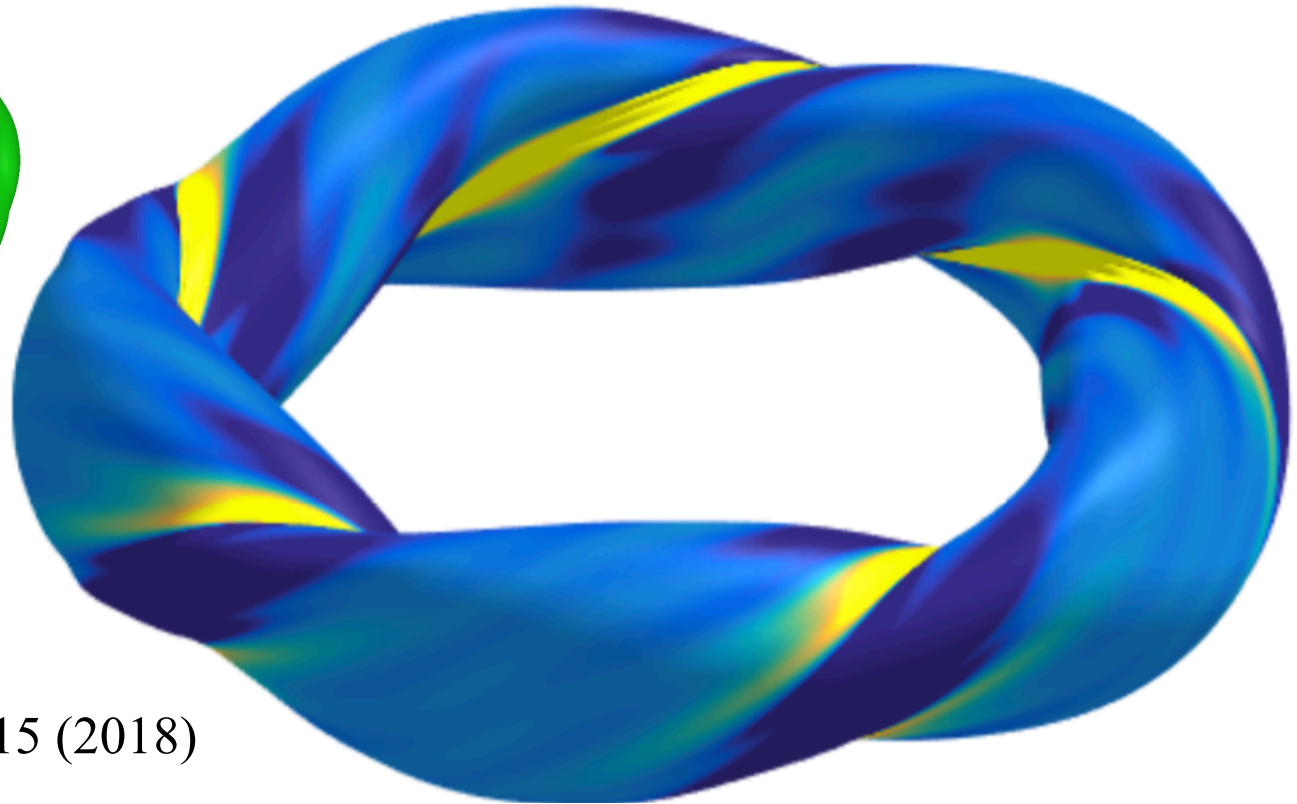
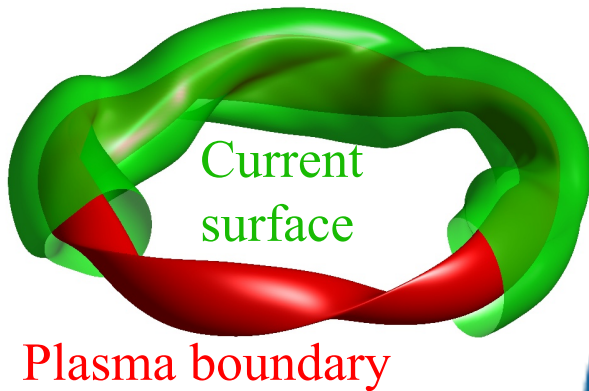
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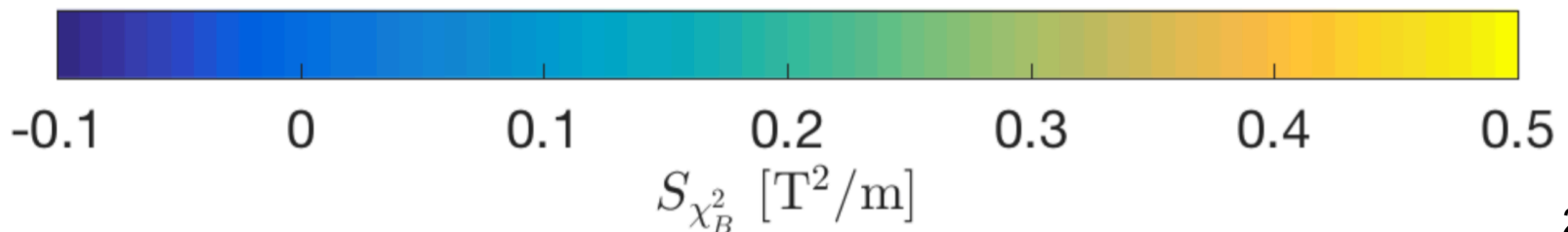


# The shape gradient can show where coils must be close to the plasma.

Shape gradient on a current surface for  $f = \int d^2a (\mathbf{B} \cdot \mathbf{n})^2$   
given a fixed plasma boundary



E. J. Paul et al,  
*Nuclear Fusion* 58, 076015 (2018)



# Understanding local sensitivity of stellarators using shape gradients

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- Coil tolerances
- Magnetic sensitivity and tolerances

# Coil tolerances can be computed from the shape gradient.

Choose an acceptable  $\Delta f$  & any weight  $w(\ell) \geq 0$ .

$$\text{Let } T(\ell) = \frac{w(\ell) \Delta f}{\sum \int d\ell' w(\ell') \|\mathbf{s}(\ell')\|}.$$

If  $|\delta \mathbf{r}| \leq T$ ,

$$|\delta f| \leq \int d\ell |\mathbf{s} \cdot \delta \mathbf{r}| \leq \int d\ell \|\mathbf{s}\| |\delta \mathbf{r}| \leq \int d\ell \|\mathbf{s}\| T = \Delta f.$$



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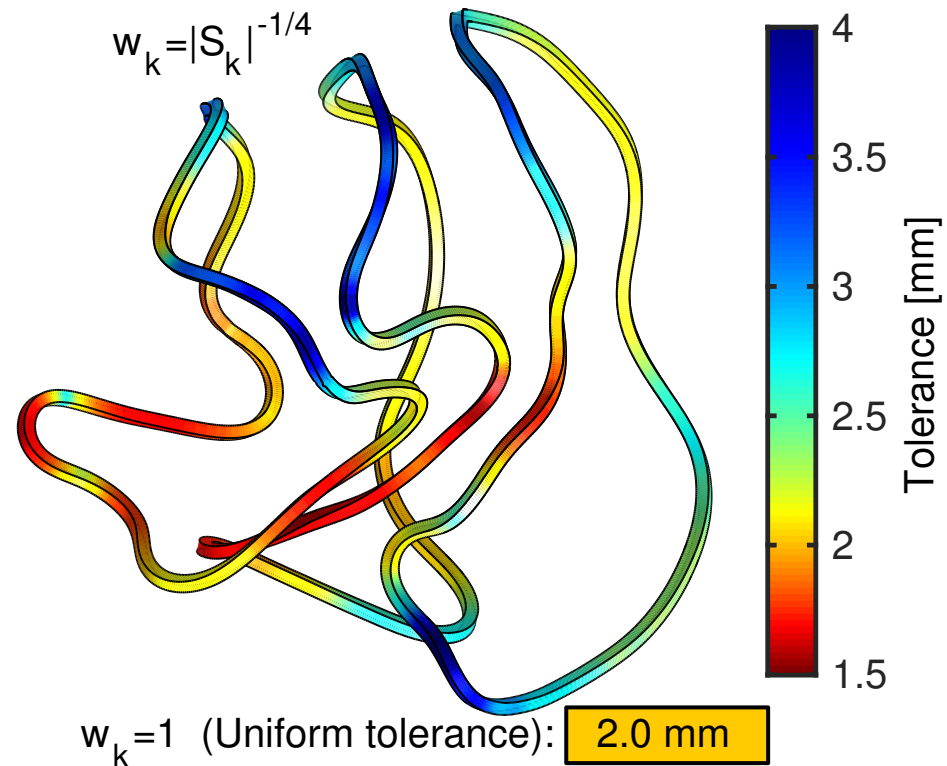
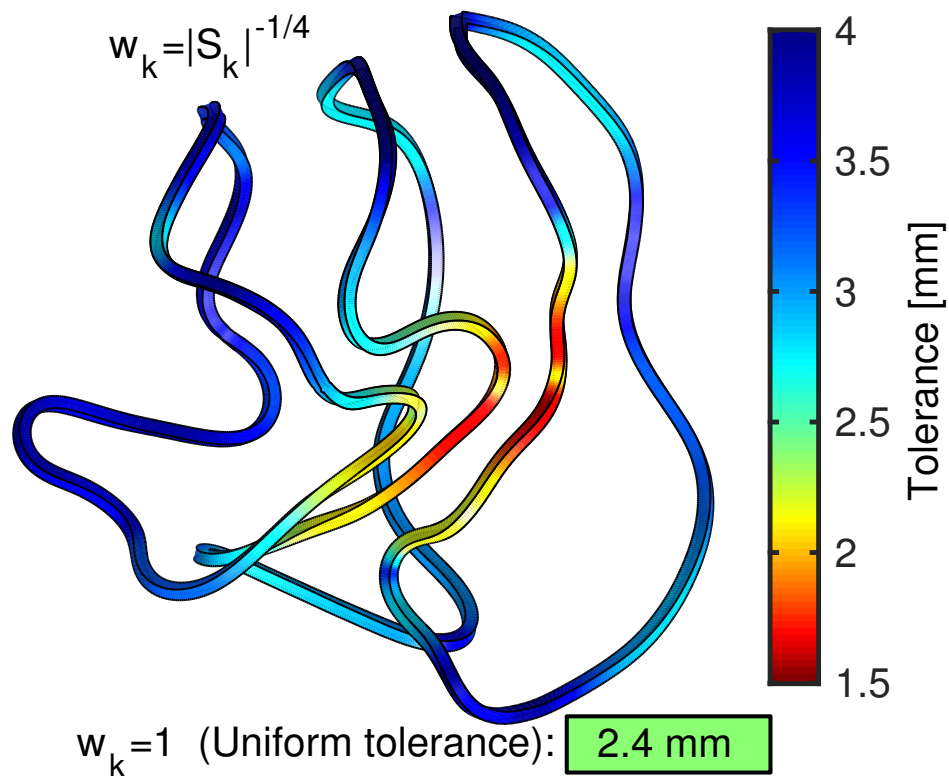
$$|\delta f| \leq \int d\ell |\mathbf{s} \cdot \delta \mathbf{r}| \leq \int d\ell \|\mathbf{s}\| |\delta \mathbf{r}| \leq \int d\ell \|\mathbf{s}\| T = \Delta f.$$

Conservative: a bound on the worst possible outcome.

# Coil tolerances can be computed from the shape gradient.

$$\Delta t = 0.02$$

$$\Delta \varepsilon_{eff}^{3/2} = \varepsilon_{eff}^{3/2} / 2$$



# Understanding local sensitivity of stellarators using shape gradients

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# A magnetic sensitivity $S_B$ can be computed from the shape gradient.

Define  $S_B$  by  $\mathbf{B}_0 \cdot \nabla S_B = \langle S \rangle - S$ .

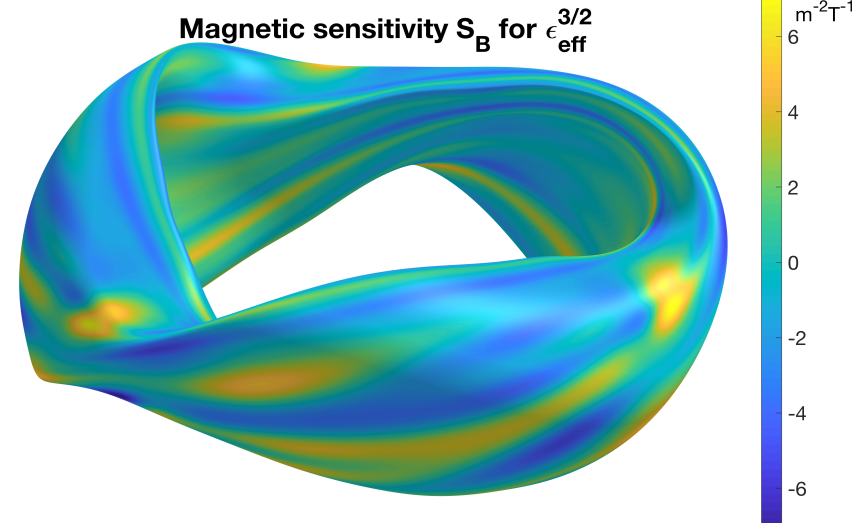
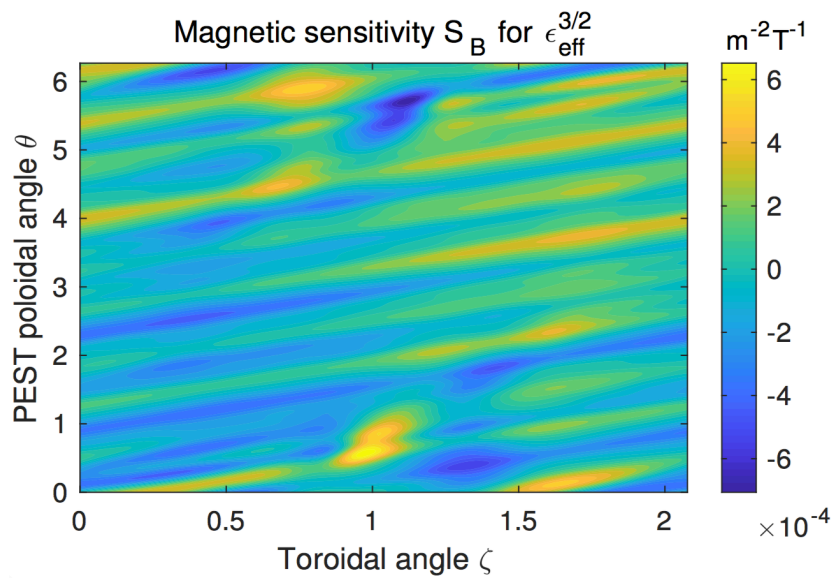
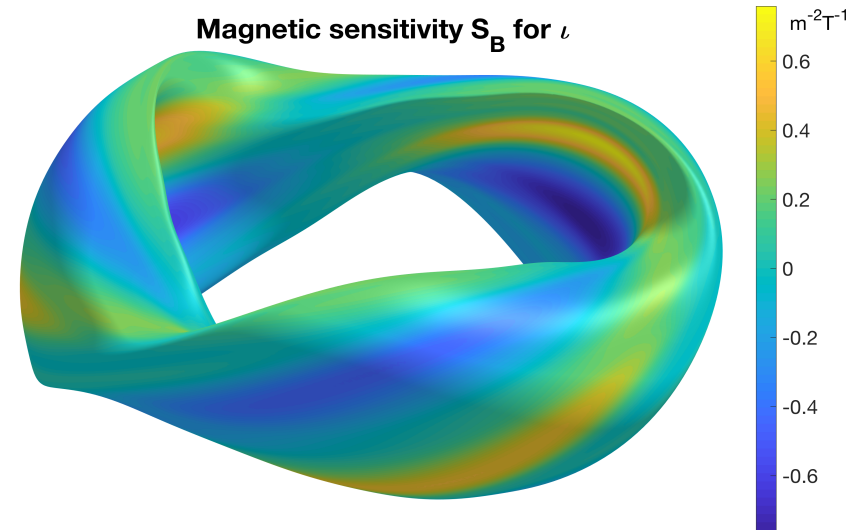
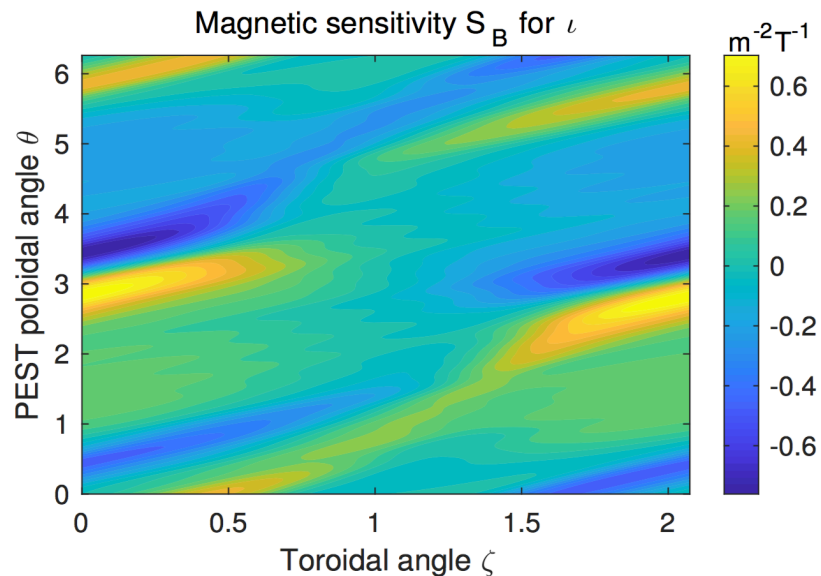
Substitute into  $\delta f = \int d^2a \, S \delta \mathbf{r} \cdot \mathbf{n}$ .

After some algebra ...

$$\Rightarrow \delta f = \int d^2a \, S_B \delta \mathbf{B} \cdot \mathbf{n}.$$

# A magnetic sensitivity $S_B$ can be computed from the shape gradient.

$$\delta f = \int d^2a S_B \delta \mathbf{B} \cdot \mathbf{n} \quad \text{where} \quad \mathbf{B}_0 \cdot \nabla S_B = \langle S \rangle - S.$$



# A magnetic tolerance $T_B$ can be computed from the magnetic sensitivity.

Choose an acceptable  $\Delta f$  & any weight  $W(\theta, \zeta) \geq 0$ .

$$\text{Let } T_B(\theta, \zeta) = \frac{W(\theta, \zeta) \Delta f}{\int d^2 a' W(\theta', \zeta') \|S_B(\theta', \zeta')\|}.$$

If  $|\delta \mathbf{B} \cdot \mathbf{n}| \leq T_B$ ,

$$\begin{aligned} |\delta f| &\leq \int d^2 a \|S_B\| |\delta \mathbf{B} \cdot \mathbf{n}| \\ &\leq \int d^2 a \|S_B\| T_B \\ &\leq \Delta f. \end{aligned}$$

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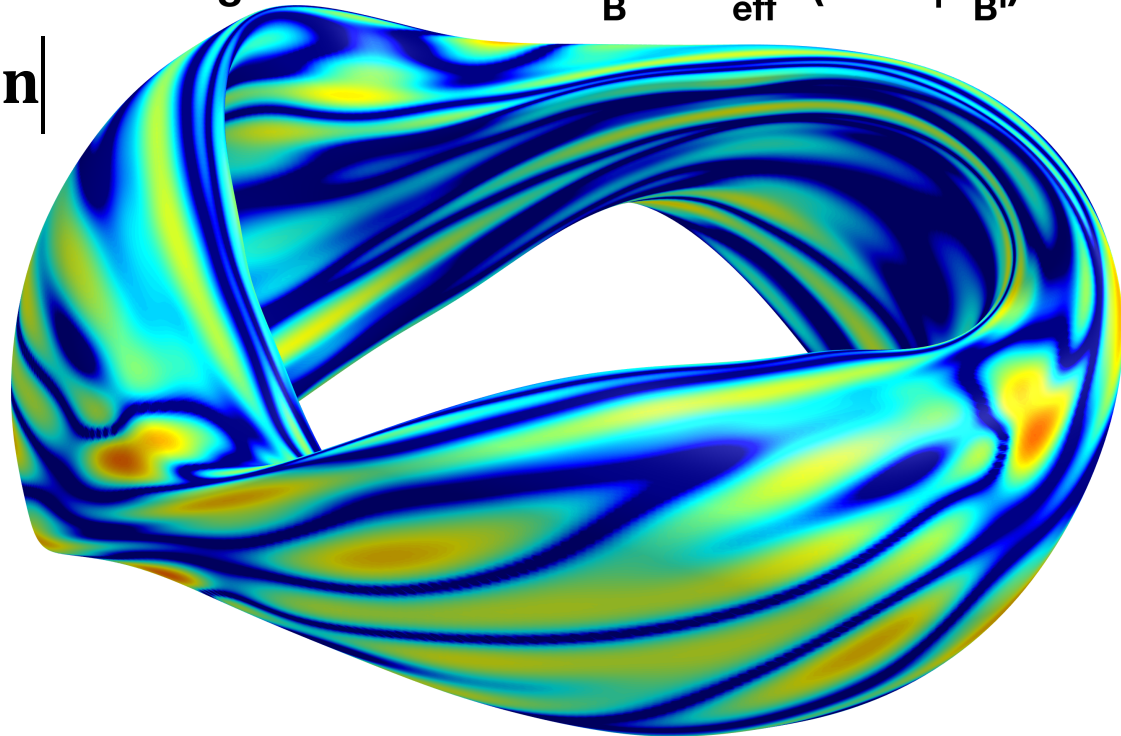
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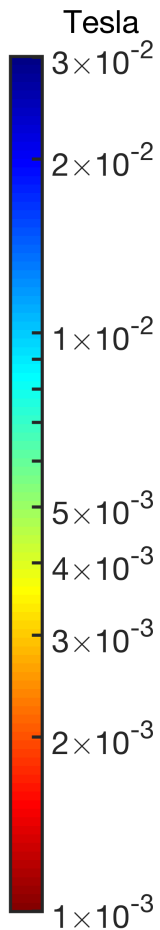
$$\begin{aligned} |\delta f| &\leq \int d^2 a |S_B| |\delta \mathbf{B} \cdot \mathbf{n}| \\ &\leq \int d^2 a |S_B| T_B \\ &\leq \Delta f. \end{aligned}$$

Magnetic tolerance  $T_B$  for  $\epsilon_{\text{eff}}^{3/2}$  ( $W=1/|S_B|$ )



$W=1$  (Alternative uniform tolerance):

$6.1 \times 10^{-3} \text{ T}$





# Conclusions

Shape gradients provide *local* sensitivity & tolerance information which can inform

- How accurately and rigidly the coils should be built,
- Where coils should be connected to support structure,
- Where sources of error fields like coil leads should be located, and enable systematic calculation of tolerances.

Future work:

- Shape gradients for integrability/islands,
- Direct calculation of shape gradients using adjoint methods,
- Target tolerances in STELLOPT to increase them.

*Nuclear Fusion 58, 076023 (2018)*

**Extra slides**

**The shape gradient can be computed from parameter derivatives by solving a small linear system.**

## Plasma boundary shape:

Parameters  $p_j$  are  $\{R_{m,n}^c, Z_{m,n}^s\}$ .

Compute  $\frac{\partial f}{\partial p_j}$  using finite differences, e.g. STELLOPT.

Discretize shape gradient:  $S(\theta, \zeta) = \sum_q S_q \cos(m_q \theta - n_q \zeta)$

**The shape gradient can be computed from parameter derivatives by solving a small linear system.**

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$$\int d^2a (\delta \mathbf{r} \cdot \mathbf{n}) S = \delta f \quad \Rightarrow \quad \text{Solve } \int d^2a \frac{\partial \mathbf{r}}{\partial p_j} \cdot \mathbf{n} S = \frac{\partial f}{\partial p_j} \text{ for } S. \quad (1)$$

Linear system, not square.

# The shape gradient can be computed from parameter derivatives by solving a small linear system.

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Compute  $\frac{\partial f}{\partial p_j}$  using finite differences, e.g. STELLOPT.

Discretize shape gradient:  $S(\theta, \zeta) = \sum_q S_q \cos(m_q \theta - n_q \zeta)$

$$\int d^2a (\delta \mathbf{r} \cdot \mathbf{n}) S = \delta f \quad \Rightarrow \quad \text{Solve } \int d^2a \frac{\partial \mathbf{r}}{\partial p_j} \cdot \mathbf{n} S = \frac{\partial f}{\partial p_j} \text{ for } S. \quad (1)$$

Linear system, not square.

Check that  $\frac{\partial f}{\partial p_j}$  is in the column space of matrix.

If so, (1) can be solved for  $S_q$  using pseudo-inverse of matrix.

# Shape gradients have been used in other fields, mostly for shapes interacting with neutral fluid.

- O. Pironneau, *Optimal Shape Design for Elliptic Systems* (Springer-Verlag, 1984).
- K. K. Choi and N.-H. Kim, *Structural Sensitivity Analysis and Optimization, vol 1* (Springer, 2005).
- M. C. Delfour and J.-P. Zolesio, *Shapes and Geometries: Metrics, Analysis, Differential Calculus, and Optimization*, 2<sup>nd</sup> ed. (SIAM, 2011).
- ...

And recently for optimizing tokamak divertor shapes:

- W. Dekeyser, Ph.D. thesis, KU Leuven (2014).
- W. Dekeyser et al, *Nucl. Fusion* 54, 073022 (2014).
- M. Baelmans, et al, *Nucl. Fusion* 57, 036022 (2017).

# The shape gradient representation can be expected to exist for many important shape functionals.

Derivative of a function of  $n$  numbers  $f(r_1, r_2, \dots, r_n)$ :  $\delta f = \sum_{j=1}^n \frac{\partial f}{\partial r_j} \delta r_j$

$$n \rightarrow \infty \text{ limit: } f = f[r(\vartheta)], \quad \delta f = \int_0^{2\pi} d\vartheta \frac{\delta f}{\delta r} \delta r$$

This is an instance of the *Riesz representation theorem*: any linear operator can be written as an inner product with some element of the appropriate space.

$$\text{3D: } f = f[r_X(\vartheta), r_Y(\vartheta), r_Z(\vartheta)], \quad \delta f = \sum_{i=X,Y,Z} \int_0^{2\pi} d\vartheta \frac{\delta f}{\delta r_i} \delta r_i$$

$$\text{Define } \mathbf{S} = \left| \frac{d\mathbf{r}}{d\vartheta} \right|^{-1} \left( \mathbf{e}_X \frac{\partial f}{\partial r_X} + \mathbf{e}_Y \frac{\partial f}{\partial r_Y} + \mathbf{e}_Z \frac{\partial f}{\partial r_Z} \right) \quad \Rightarrow \quad \delta f = \int d\ell \mathbf{S} \cdot \delta \mathbf{r}$$



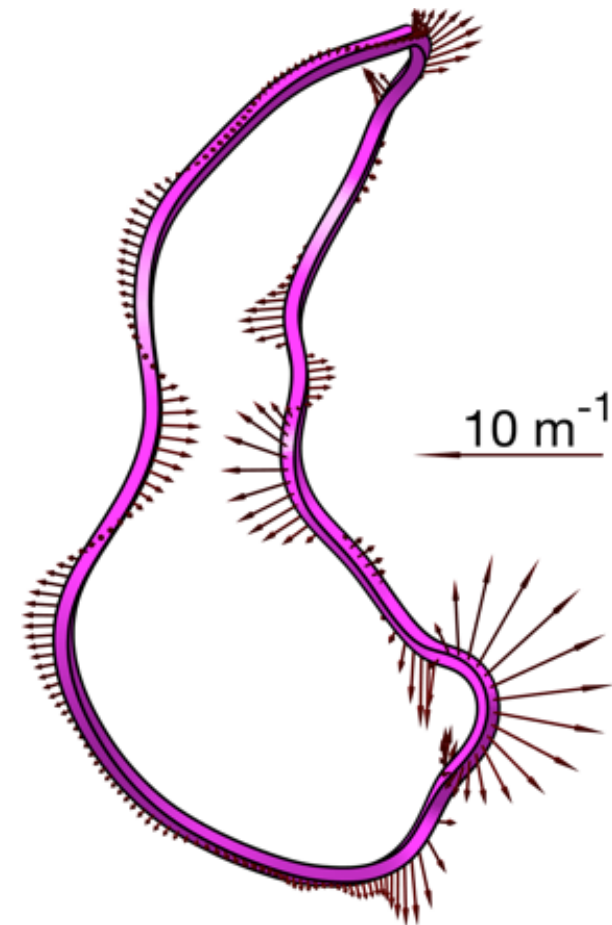
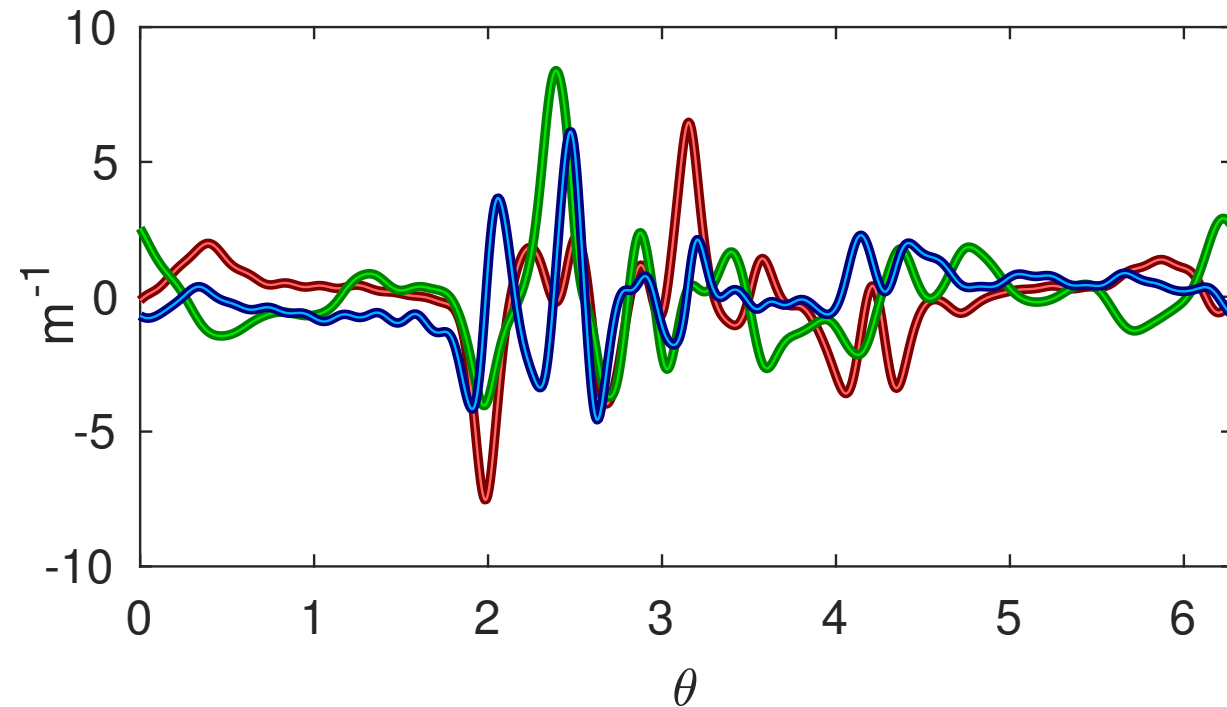
# The algorithm for computing coil shape gradients can be verified by comparison to analytic theory.

Consider  $f = \text{length}$ . Analytic result:  $\mathbf{S} = -\kappa \mathbf{n}$

(NCSX type-A coil)

Shape gradient  $\mathbf{S}$ :

Finite diff. method:  $\text{--- } S_X$   $\text{--- } S_Y$   $\text{--- } S_Z$   
Analytic answer:  $\text{--- } -\kappa n_X$   $\text{--- } -\kappa n_Y$   $\text{--- } -\kappa n_Z$



# 'Direct method' to compute sensitivity map

$$\int d^2a S_d \frac{\partial \mathbf{r}}{\partial p_k} \cdot \mathbf{n} = \frac{\partial f}{\partial p_k} \quad p_k \in \{R_{mn}^c, Z_{mn}^s\}$$

Unknowns =  $S_d$  on a grid:  $S_d(\theta_j, \zeta_j)$

$$\underbrace{\left( \int d^2a \frac{\partial \mathbf{r}}{\partial p_k} \cdot \mathbf{n} \right)}_{\mathbf{D}} \underbrace{\left( S_d \right)}_{\mathbf{S}} = \underbrace{\left( \frac{\partial f}{\partial p_k} \right)}_{\mathbf{j}}$$

SVD:  $\mathbf{D} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$       Solution by pseudo-inverse:  $\mathbf{S} = \mathbf{V}\mathbf{\Sigma}^+ \mathbf{U}^T \mathbf{j}$

If  $\mathbf{j}$  is not in the column space of  $\mathbf{D}$ , then no sensitivity map exists.

$\Rightarrow$  Check whether  $\mathbf{U}^T \mathbf{j}$  entries are small where  $\mathbf{\Sigma}$  entries are small.

# Fourier method to compute sensitivity map

$$\int d^2a S_d \frac{\partial \mathbf{r}}{\partial p_k} \cdot \mathbf{n} = \frac{\partial f}{\partial p_k} \quad p_k \in \{R_{mn}^c, Z_{mn}^s\} \quad S_d = \sum_{m,n} S_{m,n} \cos(m\theta - n\zeta)$$

Unknowns:  $S_{m,n}$

$$\underbrace{\left( \int d^2a \left( \frac{\partial \mathbf{r}}{\partial p_k} \cdot \mathbf{n} \times \cos(m\theta - n\zeta) \right) \right)}_{\mathbf{D}} \underbrace{\left( S_{m,n} \right)}_{\mathbf{S}} = \underbrace{\left( \frac{\partial f}{\partial p_k} \right)}_{\mathbf{j}}$$

SVD:  $\mathbf{D} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$       Solution by pseudo-inverse:  $\mathbf{S} = \mathbf{V}\mathbf{\Sigma}^+ \mathbf{U}^T \mathbf{j}$

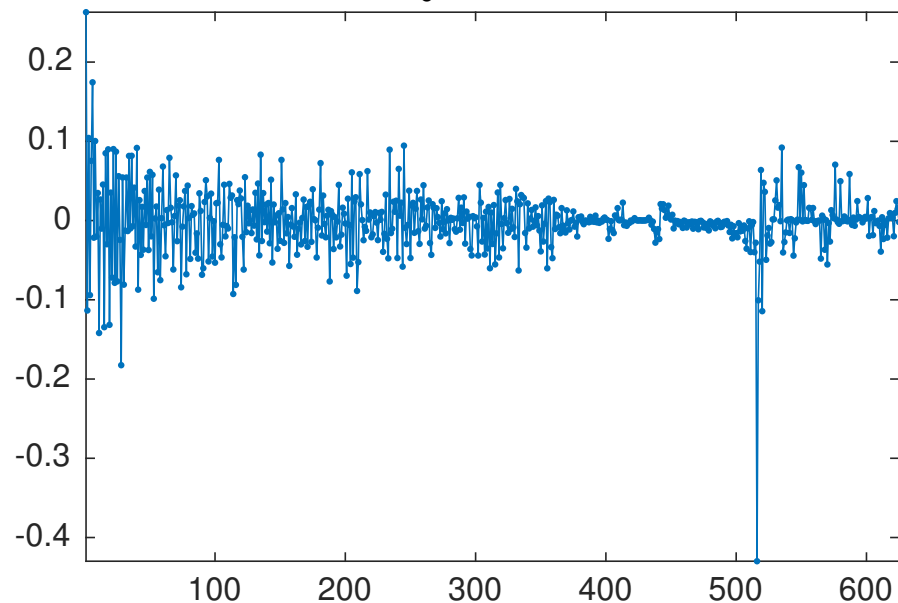
If  $\mathbf{j}$  is not in the column space of  $\mathbf{D}$ , then no sensitivity map exists.

$\Rightarrow$  Check whether  $\mathbf{U}^T \mathbf{j}$  entries are small where  $\mathbf{\Sigma}$  entries are small.

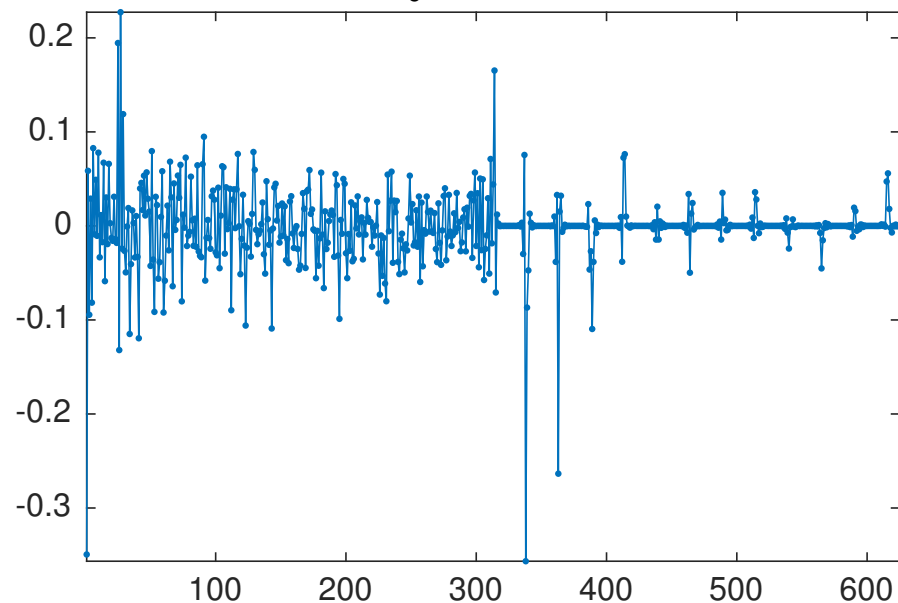
**If  $f$  is coordinate-dependent, so a sensitivity map does not exist, the RHS is not in the column space.**

$$\text{Example: } f = R_0 = \frac{1}{(2\pi)^2} \int_0^{2\pi} d\theta \int_0^{2\pi} d\zeta R$$

(a)  $U^T j$  for  $R_0$ , direct approach



(b)  $U^T j$  for  $R_0$ , Fourier approach



# A magnetic sensitivity map can be computed from the displacement sensitivity map.

$$\mathbf{B} = \mathbf{B}_0 + \delta\mathbf{B}, \quad \psi = \psi_0 + \delta\psi \quad \mathbf{B} \cdot \nabla \psi = 0 \quad \mathbf{B}_0 \cdot \nabla \delta\psi + \delta\mathbf{B} \cdot \nabla \psi_0 = 0$$

$$0 = d\psi = \delta\psi + \delta\mathbf{r} \cdot \nabla \psi_0 \quad \Rightarrow \quad \delta\psi = -\delta\mathbf{r} \cdot \nabla \psi_0$$

$$\mathbf{B}_0 \cdot \nabla (\delta\mathbf{r} \cdot \nabla \psi_0) = \delta\mathbf{B} \cdot \nabla \psi_0$$

Equivalent to the  $\nabla \psi_0$  component  
of the MHD induction equation  $\delta\mathbf{B} = \nabla \times (\delta\xi \times \mathbf{B}_0)$ .

Define  $S_B$  by  $\mathbf{B}_0 \cdot \nabla S_B = \langle S_d \rangle - S_d$ .

$$\begin{aligned} \delta f = \int d^2a S_s \delta\mathbf{r} \cdot \mathbf{n} &= \langle S_d \rangle \int d^2a \delta\mathbf{r} \cdot \mathbf{n} - \int d^2a \delta\mathbf{r} \cdot \mathbf{n} \mathbf{B}_0 \cdot \nabla S_B \\ &= \langle S_d \rangle \delta V + \int d^2a \frac{S_B}{|\nabla \psi_0|} \mathbf{B}_0 \cdot \nabla (\delta\mathbf{r} \cdot \nabla \psi_0) \end{aligned}$$

By parts

$$\delta f = \langle S_d \rangle \delta V + \int d^2a S_B \delta\mathbf{B} \cdot \mathbf{n}$$